

Proportions (z)

One sample

sample is randomly selected from population of ____ ; sample size is < 10% of population size
 sample size is large enough: **hyp test:** $np \geq 10$ and $n(1-p) \geq 10$ **Conf Int:** $n\hat{p} \geq 10$ and $n(1-\hat{p}) \geq 10$

So I will use z procedures for a population proportion.

hyp test:

$$z = \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}}$$

conf interval:

$$\hat{p} \pm z^* \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

Two samples

both samples are randomly selected and selected independently

both sample sizes large enough $n_1\hat{p}_1 \geq 10$ $n_1\hat{q}_1 \geq 10$ $n_2\hat{p}_2 \geq 10$ $n_2\hat{q}_2 \geq 10$

So I will use z procedures for the difference of two population proportions.

$$z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{p_c(1-p_c)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

$$p_c = \frac{\text{Total Successes}}{\text{Total Samples}} = \frac{n_1\hat{p}_1 + n_2\hat{p}_2}{n_1 + n_2} = \frac{x_1 + x_2}{n_1 + n_2}$$

$$\hat{p}_1 - \hat{p}_2 \pm z^* \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

Chi-Square (Counts) $\chi^2 = \sum \frac{(Obs - Exp)^2}{Exp}$

	<u>goodness-of-fit</u>	<u>homogeneity</u>	<u>independence</u>
hypothesis	Is sample prop diff from hypothesized population distribution? H_O : they are same (be specific) H_A : they are different	proportions of 2 or more pop are different H_O : $p_1 = p_2 = p_3$ H_A : they are different	two variables are independent (no association) H_O : no association between H_A : an association between ...
assumptions	all <u>expected</u> counts ≥ 5 random sample	all <u>expected</u> counts ≥ 5 random, independent samples	all <u>expected</u> counts ≥ 5 random sample
	So, I will use chi-square procedures for goodness of fit test with $k-1$ degrees of freedom ($k = \#$ of categories)	So, I will use chi-square procedures for test for homogeneity with $(r-1)(c-1)$ df	So, I will use chi-square procedures for test for independence with $(r-1)(c-1)$ df

MEANS (z or t)

One sample (st dev of population is known)

sample is randomly selected or assigned
 population is fairly normally dist
 OR sample size is large enough ($n > 30$)
 OR boxplot of sample shows no outliers, so plausible that pop is normal
So, I will use z procedures for pop means

$$z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \quad \bar{x} \pm z * \frac{\sigma}{\sqrt{n}}$$

One sample (st dev of population is NOT KNOWN) use t

sample is randomly selected or assigned
 population is fairly normally dist
 OR sample size is large enough ($n > 30$)
 OR boxplot of sample shows no outliers, so plausible that pop is normal
So, I will use t procedures for means with n-1 df

$$t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \quad \bar{x} \pm t * \frac{s}{\sqrt{n}}$$

Two samples Check **both** samples for:
 samples are independent, random from _____
 st dev of populations are **known**
 population is normal OR
 sample sizes are large ($n \geq 30$)
So, I will use z procedures for difference of 2 population means

Matched pairs

- matched (state how);
- sample randomly selected from pop of differences
- large ($n \geq 30$) OR boxplot of single list of differences shows no outliers
 (one list of differences is created from two matched lists)

So I will use t procedures for matched pairs (one sample of differences)
n-1 df formulas: see above-right for one sample t

Two samples: Check **both** samples for:
 samples are random, independent (except treatments)
 st dev of populations are **not known** use t
 populations are normal OR
 sample sizes are large (both $n \geq 30$) OR boxplots of samples
 show no outliers, so plausible that both pops are normal

**So, I will use t procedures for difference of 2 population means
 using smaller n-1 degrees of freedom**

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - 0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \quad (\bar{x}_1 - \bar{x}_2) \pm t * \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

t test for slope or model utility test or linear regression t test

hypothesis

$$H_0: \beta = 0$$

$$H_a: \beta \neq 0$$

conditions

scatterplot of the data shows a linear pattern
 variability of residuals does not appear to be changing with x

We hope that the residuals from the line at any given x are normally
 distributed with mean = 0

(check histogram or normal probability plot of residuals)

$$t = \frac{b_1}{s_{b_1}}$$

$$b_1 \pm t * s_{b_1}$$

So, I will use the linear regression t procedure with df = n - 2